

Newton's Polynomial: **General Form** (n^{th} order)

- Note: Calculating these divided differences is a **recursive** process:

$$f_{dd}[x_3, x_2, x_1, x_0]$$

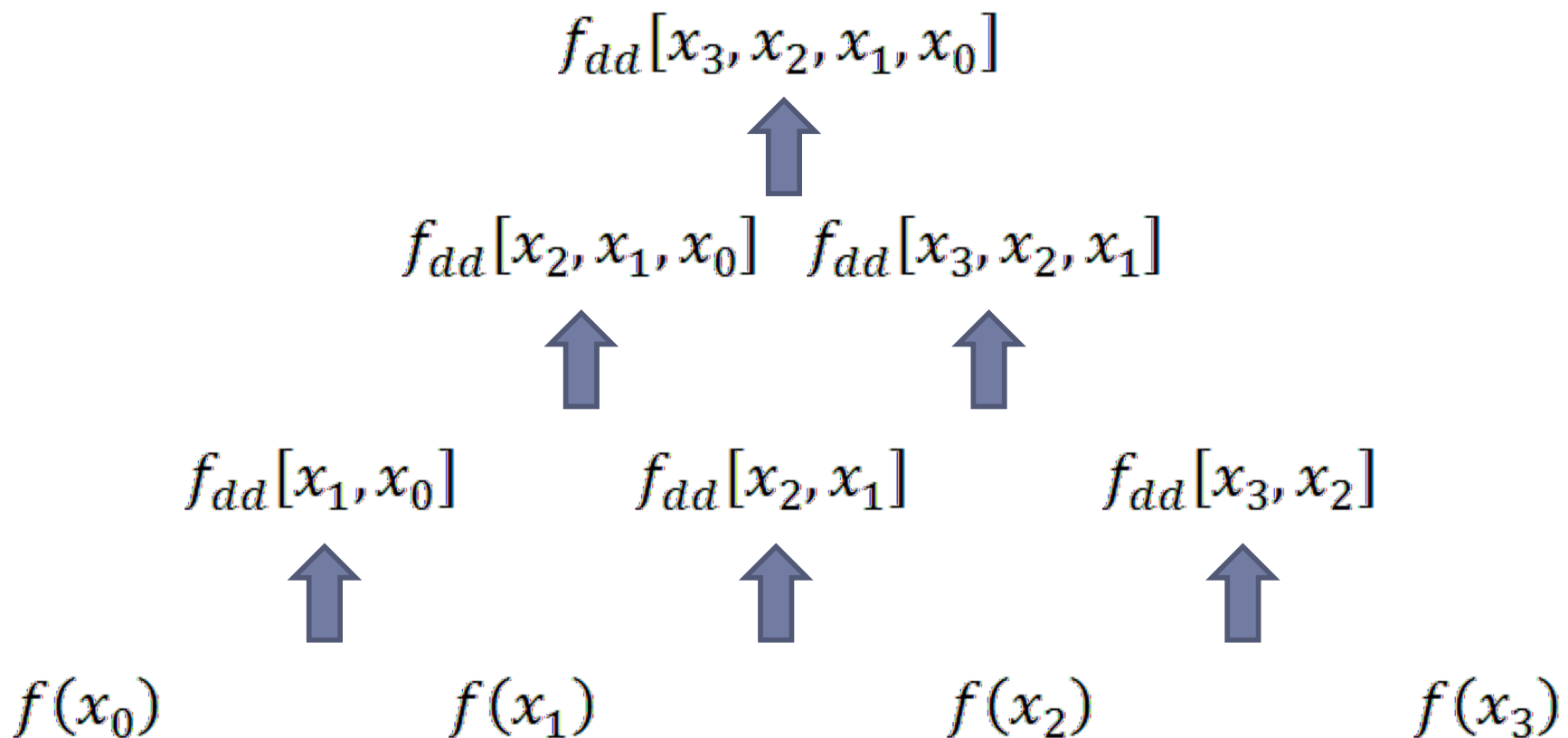
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$$f_{dd}[x_1, x_0] \quad f_{dd}[x_2, x_1] \quad f_{dd}[x_3, x_2]$$

$$f(x_0) \quad f(x_1) \quad f(x_2) \quad f(x_3)$$

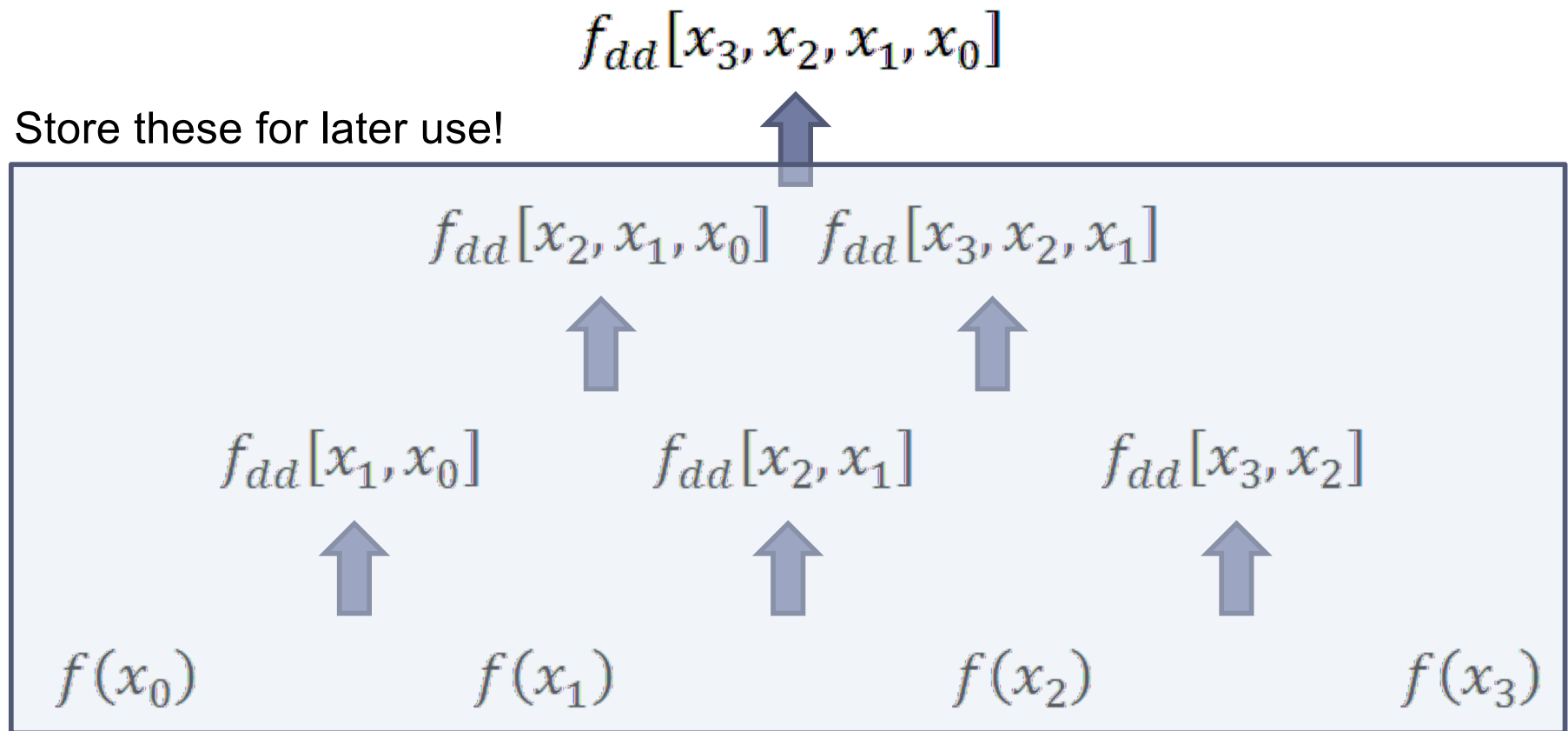
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Newton's Polynomial: **General Form (nth order)**

- Putting these divided differences into the polynomial form:

$$\begin{aligned} f_3(x) = & f(x_0) \\ & + f_{dd}[x_1, x_0](x - x_0) \\ & + f_{dd}[x_2, x_1, x_0](x - x_0)(x - x_1) \\ & + f_{dd}[x_3, x_2, x_1, x_0](x - x_0)(x - x_1)(x - x_2) \end{aligned}$$

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- This is the general form of Newton's divided-difference interpolating polynomial

Newton's Polynomial: Example 18.1 & 18.2 – $\ln(x)$

- Estimate $f(2)$ using 3rd order polynomial:

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1) \\ + b_3(x - x_0)(x - x_1)(x - x_2)$$

- Need 4 points, so **add one more**

	x	$f(x)$
x_0	1	0
x_1	4	1.386294
x_2	6	1.791759
x_3	5	1.609438

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 - Note that points need be sequential

- Recall:

$$b_0 = f(x_0)$$

$$b_1 = f_{dd}[x_1, x_0]$$

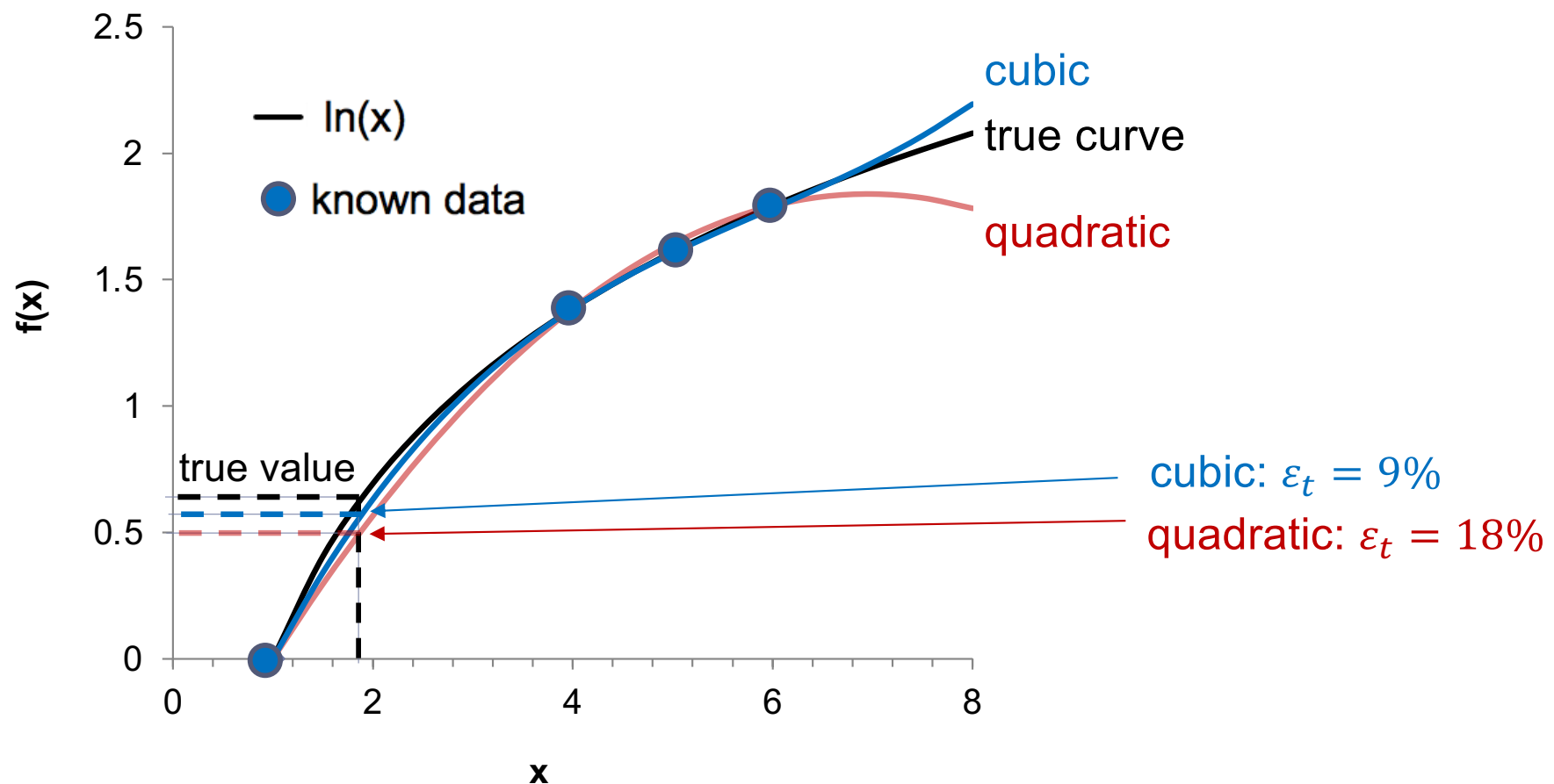
$$b_2 = f_{dd}[x_2, x_1, x_0]$$

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MIE334_Lecture_21_ExNewt3.xlsx

Newton's Polynomial: Example 18.3 - Results

- Higher order further reduces the true error



Newton's Polynomial: **Error**

- General form of this polynomial looks like a Taylor series:

$$\begin{aligned} f_n(x) = & f(x_0) \\ & + f_{dd}[x_1, x_0](x - x_0) \\ & + f_{dd}[x_2, x_1, x_0](x - x_0)(x - x_1) \\ & \quad \vdots \quad \quad \quad \vdots \\ & + f_{dd}[x_n, x_{n-1}, \dots, x_0](x - x_0)(x - x_1) \cdots (x - x_{n-1}) \end{aligned}$$

- Successive terms are added to estimate higher-order behavior (curvature) of the function

Newton's Polynomial: Error

- We can derive an estimate of the error in our interpolation using a property of the Taylor series
- Recall: Truncation error for the Taylor series:
 - $R_n = \frac{f^{(n+1)}(\xi)}{(n+1)!} \overbrace{(x_{i+1} - x_i)^{n+1}}^h$
 - ξ is somewhere (unknown) between x_{i+1} and x_i
- For a n^{th} -order polynomial interpolation we can use an analogous term to replace the $(x_{i+1} - x_i)^{n+1}$ term:
 - $(x - x_0)(x - x_1) \dots (x - x_n)$
 - Still a polynomial of order $n + 1$

Newton's Polynomial: **Error**

- So error term now looks like:
 - $R_n = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x - x_0)(x - x_1) \dots (x - x_n)$
 - ξ lies somewhere in range of data points
- We can estimate the $(n+1)$ derivative using a $(n+1)$ finite divided difference:
 - $\frac{f^{(n+1)}(\xi)}{(n+1)!} \cong f_{dd}[x_{n+1}, x_n, x_{n-1}, \dots, x_0]$
- To estimate the error, we need one **additional data point**:
 - $R_n = f_{dd}[\mathbf{x_{n+1}}, x_n, x_{n-1}, \dots, x_0](x - x_0)(x - x_1) \dots (x - x_n)$

Newton's Polynomial: Error - Example

- Earlier, we estimated $f(2)$ with a 2nd-order polynomial

- Original data points are the same
- But use additional point, $f(x_3) = f(5)$, to estimate the error for this interpolation

x	$f(x)$
1	0
4	1.386294
6	1.791759
5	1.609438

- From the 2nd order fit we found:

- $f_2(2) = 0.565844$, $\varepsilon_t = 18.4\%$

- $$R_2 = f_{dd}[x_3, x_2, x_1, x_0](x - x_0)(x - x_1)(x - x_2)$$
$$= f_{dd}[5, 6, 4, 1](2 - 1)(2 - 4)(2 - 6) = \mathbf{0.062924}$$

- Same order of magnitude as true error ($E_t = 0.127303$) but about half the value